

### 8.3: Hypothesis Tests in Detail

**Definition. (Type I and type II errors)**

- A **type I error** is rejecting the null hypothesis,  $H_0$ , when it is actually true.
- A **type II error** is failing to reject the null hypothesis,  $H_0$ , when it is false.

The probability of committing a type I error is the level of significance:  $\alpha$



	Null Hypothesis is true	Null Hypothesis is false
Reject null hypothesis	Type I error	True positive
Fail to reject null hypothesis	True negative	Type II error

**Example.** For the following scenarios, identify the type I and type II errors:

“The Boy Who Cried Wolf”

In a court of law, a person is considered innocent until proven guilty.

Testing someone for a disease (e.g. Covid)

Pregnancy test

- When we “fail to reject  $H_0$ ”, we are **not proving** the null hypothesis
- Don’t change your hypothesis after you gather your results
- Statistically significant means something likely did not occur by chance
- Confidence intervals vs. Hypothesis testing

Confidence Intervals	Hypothesis Tests
Estimates parameters	Test parameters
Range of values	Is data consistent?

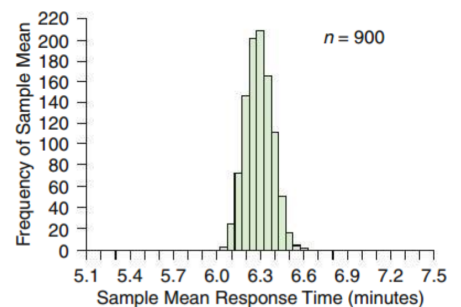
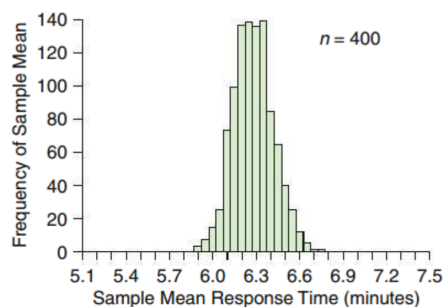
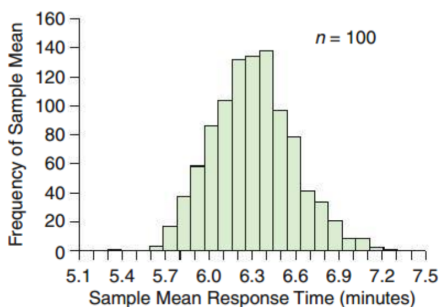
## 9.1: Sample Means of Random Samples

Statistics	Parameters		
Sample mean	$\bar{x}$	Population mean	$\mu$
Sample standard deviation	$s$	Population standard deviation	$\sigma$
Sample proportion	$\hat{p}$	Population proportion	$p$

### Definition.

- The **sampling distribution** is the distribution of the sample means  $\bar{x}$ .
- The mean of the sampling distribution is  $\mu$  so the statistic  $\bar{x}$  is an **unbiased estimator**.
- The standard deviation of the sampling distribution is the **standard error**:

$$SE = \frac{\sigma}{\sqrt{n}}$$



## 9.2: The Central Limit Theorem for Sample Means

### Definition. (Central Limit Theorem (CLT))

When estimating a population mean,  $\mu$ , if

1. *Random and Independent*: Each observation is collected randomly from the population, and observations are independent of each other.
2. *Large Sample*: Either the population distribution is Normal, or the sample size is large ( $n \geq 25$ ).
3. *Big population*: If the sample is collected without replacement (e.g. SRS), then the population size must be at least 10 times bigger than the sample size.

$$N \geq 10n$$

then the sampling distribution for  $\bar{x}$  is approximately Normal, with mean  $\mu$  and standard deviation

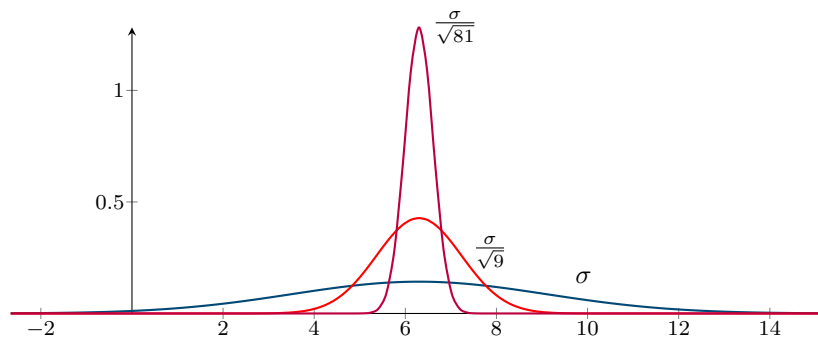
$$SE = \frac{\sigma}{\sqrt{n}}.$$

This distribution is denoted as

$$N\left(\mu, \frac{\sigma}{\sqrt{n}}\right).$$

**Example.** The population distribution of *all* emergency response times from the LA Fire Department is right-skewed. Suppose we repeatedly take random samples of a certain size from this population and calculate the mean response time. We know that the population has mean  $\mu = 6.3$  and standard deviation  $\sigma = 2.8$  minutes.

Describe the sampling distribution if the sample size is  $n = 9$ , and again when  $n = 81$ .



**Note:** Even if the population distribution has an unusual shape, the sampling distribution is fairly symmetric and unimodal.

**Example.** According to one very large study done in the US, the mean resting pulse rate of adult women is about  $\mu = 74$  BPM, with standard deviation  $\sigma = 13$  BPM, where the distribution is known to be skewed right. Suppose we take a random sample of 36 women from this population.

What is the approximate probability that the average pulse rate of this sample will be below 71 or above 77?

Can we find the probability that a single adult woman, randomly selected from this population, will have a resting pulse rate more than 3 BPM away from the mean value,  $\mu = 74$ ?

**Definition. (The  $t$ -Distribution)**

The hypothesis tests and confidence intervals we will use for estimating and testing the mean are based on the  **$t$ -statistic**:

$$t = \frac{\bar{x} - \mu}{SE_{est}}$$
$$SE_{est} = \frac{s}{\sqrt{n}}$$

The  $t$ -statistic follows the  **$t$ -distribution**. With the  $t$ -distribution, we do not need to check conditions for the CLT, but the distribution's shape is dependent on the **degrees of freedom (df)**.

If we know the population standard deviation  $\sigma$ , then we have the familiar  $z$ -statistic:

$$z = \frac{\bar{x} - \mu}{\left(\frac{\sigma}{\sqrt{n}}\right)}$$